

# Manual && Reproduction Report for

Watch and Act: Multi-orientation Open-set Scene Text Recognition via Dynamic Expert Routing

## Hardware requirements

GPU:

Testing: NVIDIA GPU with Turing or later structure and more than 4 Gib of vram.

(yes pascals, maxwells or even fermis may work, but you gonna need to know what you are doing)

Training: 24GiB GPU with Turing or later structure for regular model (dunno if the large one fits)

CPU: X86 cpus with AVX2 instruction set

(CPUs without AVX2 may or may not work with different version of torch, I don't know)

RAM: 16GiB (Testing) 32Gib (Training)

Disk: 200GiB

Internet: Yes

## Software Requirements

1. Fresh installed Archlinux with these packages:

`sudo vim plasma-meta nvidia openssh firefox dolphin konsole wget git less`

2. The user name is set to lasercat and it has to be a sudoer (or you may need to go thorough the code to replace paths if something went south).

3. ***Keep important data off this device!***

I don't want to wipe your data due to one or two failed cd commands followed by mv and/or rm

Trivia: Manul, aka the Pallas' cat, is the oldest cat specie still alive on the earth.



# Environment setup

1. install the following packages once you boot in.

```
sudo pacman -Syu pycharm-community-edition plasma-meta python-paramiko python-lmdb  
python-numba python-opencv python-pillow python-pip python-pyqtgraph python-pytorch-opt-  
cuda python-regex python-scikit-learn python-scipy python-torchvision-cuda python-tqdm python-  
xmldict make gcc cmake unzip python-setuptools
```

2. make dirs

```
mkdir ~/cat ~/ssddata ~/hydra_saves ~/ssddata/anchors
```

3. setup python stuff

```
mkdir ~/catvenv/; python3 -m venv ${PWD}/catvenv --system-site-packages;
```

```
source ~/catvenv/bin/activate; pip install easydict editdistance wandb
```

```
cd ~/cat; git clone https://github.com/lancercat/make_envNG.git
```

```
cd make_envNG/ ; sh pylcs.sh ;
```

```
unzip pytorch_scatter-laser.zip ; cd pytorch_scatter-2.1.2/; python setup.py install
```

4. unzip data and model

```
sh unzip.sh ${DOWNLOADPATH}
```

5. clone the code

```
cd ~/cat; git clone https://github.com/lancercat/wna.git
```

6. stop and check:

```
(catvenv) [lancercat@TESTMEOW ~]$ ls cat  
make_envNG wna  
(catvenv) [lancercat@TESTMEOW ~]$ ls hydra_saves  
aroute_nd_only-v6S-reg-nedmix-AAF-ohem01E-promelas_b32  
aroute_nd_only-v6S-reg-nedmix-AAF-ohem01E-promelas_mjst_b32  
aroute_nd_only-v6S-tiny-AAF  
aroute_nd_only-v6S-tiny-AAF-01E  
aroute_nd_only-v6S-tiny-AAF-01E-run2  
aroute_nd_only-v6S-tiny-AAF-run2  
aroute_nd_only-v6S-tiny-nedmix-AAF  
aroute_nd_only-v6S-tiny-nedmix-AAF-ohem01  
aroute_nd_only-v6S-tiny-nedmix-AAF-ohem01E  
aroute_nd_only-v6S-tiny-nedmix-AAF-ohem01E-run2  
aroute_nd_only-v6S-tiny-nedmix-AAF-ohem01-run2  
aroute_nd_only-v6S-tiny-nedmix-AAF-run2  
(catvenv) [lancercat@TESTMEOW ~]$ ls ssddata/  
anchors          ctwch          CVPR2016      IC13_1015      mltrrchlat_seen      mltrkr_hori      rctwtrdb_seen_NG  
artdb_seen       ctwdb_seen     dicts         IIIT5k_3000    mltrrchlat_seen_NG   NIPS2014         SVT  
artdb_seen_NG    ctwdb_seen_NG  dictsv2       lsvtdb_seen    mltrjp_hori         pami_ch_fsl_hwdb  
athenaNG         CUTE80         IC03_867      lsvtdb_seen_NG mltrjp_hv          rctwtrdb_seen
```



# Reproducing ablative studies

1. run `ablative.py` (in a screen session as it takes some time):

```
python ablative.py 2>&1 | tee all.log
```

(our log is uploaded to github)

2. After the ablative experiments are finished (usually takes like a dozen hours on an GTX 1650), run `ablative2table.py` to compute mean and standard deviation:

```
/home/lasercat/catvenv/bin/python3.13 /home/lasercat/cat/wna/ablative2table.py
\newcommand{\LPAoBASEoJPNHVo6ZSLoLAoAVG}{40.97}
\newcommand{\LPAoBASEoJPNHVo6ZSLoLAoSTD}{1.02}
\newcommand{\LPAoW0ARoUeJPNHVo6ZSLoLAoAVG}{42.11}
\newcommand{\LPAoW0ARoUeJPNHVo6ZSLoLAoSTD}{0.51}
\newcommand{\LPAoPGoLoJPNHVo6ZSLoLAoAVG}{41.49}
\newcommand{\LPAoPGoLoJPNHVo6ZSLoLAoSTD}{1.01}
\newcommand{\LPAoPGoLoJPNHVo6ZSLoLAoAVG}{43.09}
\newcommand{\LPAoPGoLoJPNHVo6ZSLoLAoSTD}{0.33}
\newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoAVG}{39.66}
\newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoSTD}{0.36}
\newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoAVG}{43.17}
\newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoSTD}{0.39}
\newcommand{\LPAoARoLoJPNHVo6ZSLoLAoAVG}{41.88}
\newcommand{\LPAoARoLoJPNHVo6ZSLoLAoSTD}{0.59}
\newcommand{\LPAoARoLoJPNHVo6ZSLoLAoAVG}{43.90}
\newcommand{\LPAoARoLoJPNHVo6ZSLoLAoSTD}{0.86}
\newcommand{\LPAoW0oHEMoJPNHVo6ZSLoLAoAVG}{42.05}
\newcommand{\LPAoW0oHEMoJPNHVo6ZSLoLAoSTD}{0.25}
\newcommand{\LPAoJPNHVo6ZSLoLAoAVG}{45.14}
\newcommand{\LPAoJPNHVo6ZSLoLAoSTD}{0.35}
\newcommand{\LPAoOHEMoJPNHVo6ZSLoLAoAVG}{44.53}
\newcommand{\LPAoOHEMoJPNHVo6ZSLoLAoSTD}{0.31}
-----
Process finished with exit code 0
```

3. save the output text as a txt file, say `reprod.txt`

(our log is can be found in this github repo as `reprod.txt`)

4. diff them:

```
diff reprod.txt paperref.txt --color
```

Expect the results to be a bit different here and there due to float error between different implementations of operators, which can be caused by different torch version, gpu structure, vram, etc (yes, torch[1] and CUDNN[2] may choose different backend on different hardware)

[1] <https://discuss.pytorch.org/t/different-machines-different-results/100126>

[2] <https://docs.nvidia.com/deeplearning/cudnn/backend/latest/developer/core-concepts.html>

```
(catvenv) [lasercat@TESTMEOW wna]$ diff reprod.txt paperref.txt --color
2c2
< \newcommand{\LPAoBASEoJPNHVo6ZSLoLAoSTD}{1.02}
---
> \newcommand{\LPAoBASEoJPNHVo6ZSLoLAoSTD}{1.03}
7c7
< \newcommand{\LPAoPGoLoJPNHVo6ZSLoLAoAVG}{43.09}
---
> \newcommand{\LPAoPGoLoJPNHVo6ZSLoLAoAVG}{43.07}
9,10c9,10
< \newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoAVG}{39.66}
< \newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoSTD}{0.36}
---
> \newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoAVG}{39.65}
> \newcommand{\LPAoPGoNoJPNHVo6ZSLoLAoSTD}{0.35}
13c13
< \newcommand{\LPAoARoLoJPNHVo6ZSLoLAoAVG}{41.88}
---
> \newcommand{\LPAoARoLoJPNHVo6ZSLoLAoAVG}{41.89}
18,20c18,20
< \newcommand{\LPAoW0oHEMoJPNHVo6ZSLoLAoSTD}{0.25}
< \newcommand{\LPAoJPNHVo6ZSLoLAoAVG}{45.14}
< \newcommand{\LPAoJPNHVo6ZSLoLAoSTD}{0.35}
---
> \newcommand{\LPAoW0oHEMoJPNHVo6ZSLoLAoSTD}{0.24}
> \newcommand{\LPAoJPNHVo6ZSLoLAoAVG}{45.15}
> \newcommand{\LPAoJPNHVo6ZSLoLAoSTD}{0.36}
22a23,24
```



## run object310-rel/project310\_v6SF\_stability-re/aroute\_nd\_only-v6S-tiny-nedmix-AAF-ohem01E

Let's map them to the table

Note the performance differs a bit on this 1650, but not much (utilization btw)





```
Run object310-rel/XL/aroute_nd_only-v6S-reg-nedmix-AAF-ohem01E_promelas_b32/
osocr_benchall.py
```

Let's map them to the table

Note the performance differs a bit on this 1650, but not much (utilization btw)



## **Reproducing Fig 1& Inference on your own data**

# Training from Scratch